

**Monte Carlo Simulations:  
Exercise Session**

Luca Pincirolli April 23<sup>rd</sup> 2021

### Exercise 1

Consider the exponential distribution with  $\lambda = 1$

- Sample  $N=1000$  values from  $f_T(t)$
- Verify whether the obtained distribution provides a good approximation of the analytic exponential distribution. To this aim, you are required to:
  - find the empirical probability density function (pdf) of the sampled value in 1
  - compare the empirical pdf found in 2A. with the analytical Weibull distribution.
- Provide an estimate  $G_N$  of  $\int_0^{+\infty} t f_T(t) dt$
- Estimate the variance of  $G_N$
- Consider the Weibull distribution:
 
$$F_T(t) = 1 - e^{-\beta t^\alpha}, \quad f_T(t) = \alpha \beta t^{\alpha-1} e^{-\beta t^\alpha}$$

with  $\alpha = 1.5, \beta = 1$

Provide a solution of point 1) to 4) for the Weibull distribution

Luca Pincirolli

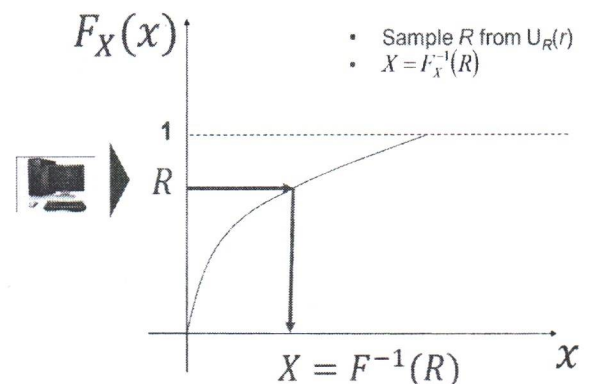
lasar

POLITECNICO MILANO 1863

### MATLAB Commands

- `rand(M,1)` provides a column of  $M$  random numbers sampled from a uniform distribution in the range  $[0,1)$
- `N = hist(Y)` bins the elements of vector  $Y$  into 10 equally spaced counters and returns the number of elements in each counter. More options if you write 'help hist'

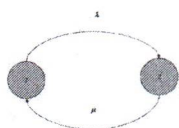
### Sampling Random Numbers from $F_X(x)$



### Exercise 2

Consider a continuously monitored component with constant failure ( $\lambda$ ) and repair ( $\mu$ ) rates in the table. Assuming a mission time  $T = 1000$  hours, write the MC code for the estimation of:

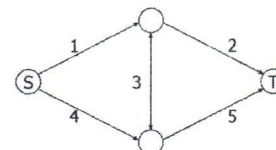
- The time dependent reliability
- The reliability at the mission time
- The instantaneous availability.



	values
$\lambda$	$3 \cdot 10^{-3} \text{ h}^{-1}$
$\mu$	$25 \cdot 10^{-3} \text{ h}^{-1}$

### Exercise 3

Consider the network in figure composed of five arcs (1, 2, 3, 4, 5). Each arc can be in two different states (1-working, 2-failed) with exponentially distributed transition times (table). The network is considered failed if there is no connection between nodes S and T. Assuming a mission time  $T_m = 300$  hours, write the MC code for the estimation of the time dependent reliability



	1	2	3	4	5
$\lambda$	$1 \cdot 10^{-3} \text{ h}^{-1}$	$2 \cdot 10^{-2} \text{ h}^{-1}$	$5 \cdot 10^{-2} \text{ h}^{-1}$	$1 \cdot 10^{-2} \text{ h}^{-1}$	$8 \cdot 10^{-4} \text{ h}^{-1}$



- The rate of transition of the system out of its current configuration (1, 1, 1) is:

$$\lambda^{(1,1,1)} = \lambda^A + \lambda^B + \lambda^C$$

- We are now in the position of sampling the first system transition time  $t_1$ , by applying the **inverse transform method**:

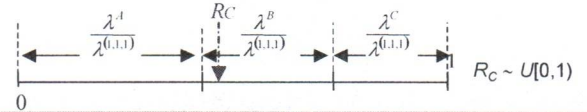
$$t_1 = t_0 - \frac{1}{\lambda^{(1,1,1)}} \ln(1 - R_1)$$

where  $R_1 \sim U[0,1)$

- Assuming that  $t_1 < T_M$  (otherwise we would proceed to the successive trial), we now need to determine which component has undergone the transition
- The probabilities of components A, B, C undergoing a transition out of their initial nominal states 1, given that a transition occurs at time  $t_1$ , are:

$$\frac{\lambda^A}{\lambda^{(1,1,1)}}, \quad \frac{\lambda^B}{\lambda^{(1,1,1)}}, \quad \frac{\lambda^C}{\lambda^{(1,1,1)}}$$

- Thus, we can apply the inverse transform method to the discrete distribution

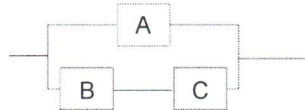


## Exercise 4

13

Consider the system in figure composed of three components(A, B, C). Each component can be in two different health states (1-nominal, 2-failed) with exponentially distributed transition times (table) between them. Assuming a mission time  $T = 500$  hours, write the MC code for the estimation of:

- The time dependent reliability
- The reliability at the mission time
- The instantaneous availability.



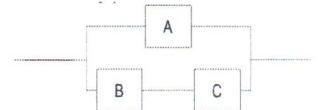
	A	B	C
$\lambda$	$1 \cdot 10^{-3} \text{ h}^{-1}$	$2 \cdot 10^{-2} \text{ h}^{-1}$	$5 \cdot 10^{-2} \text{ h}^{-1}$
$\mu$	$3 \cdot 10^{-2} \text{ h}^{-1}$	$5 \cdot 10^{-2} \text{ h}^{-1}$	$5 \cdot 10^{-3} \text{ h}^{-1}$

## Exercise 5

15

Consider the system in figure composed of three components(A, B, C). Each component can be in three different health states (1-nominal, 2-degraded, 3-failed) with exponentially distributed transition times. Assuming a mission time  $T = 1000$  hours, write the MC code for the estimation of:

- The time dependent reliability
- The reliability at the mission time
- The instantaneous availability.



Arrival	1	2	3
Initial	1	2	3
1(nominal)	0	$\lambda_{1 \rightarrow 2}^{A(B,C)}$	$\lambda_{1 \rightarrow 3}^{A(B,C)}$
2(degraded)	0	0	$\lambda_{2 \rightarrow 3}^{A(B,C)}$
3(failed)	$\lambda_{3 \rightarrow 1}^{A(B,C)}$	$\lambda_{3 \rightarrow 2}^{A(B,C)}$	0

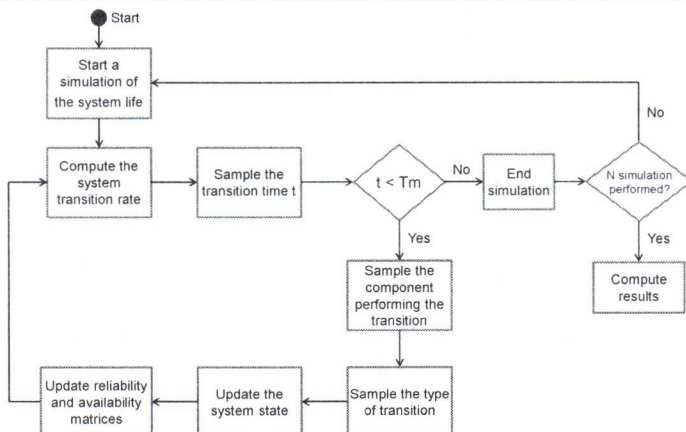
A	1	2	3
1	-	$3 \cdot 10^{-1}$	$10^{-2}$
2	-	-	$6 \cdot 10^{-1}$
3	$8 \cdot 10^{-1}$	$5 \cdot 10^{-1}$	-

B	1	2	3
1	-	$1 \cdot 10^{-1}$	$5 \cdot 10^{-1}$
2	-	-	$4 \cdot 10^{-1}$
3	$7.5 \cdot 10^{-1}$	$3.5 \cdot 10^{-1}$	-

C	1	2	3
1	-	$8 \cdot 10^{-1}$	$2.5 \cdot 10^{-1}$
2	-	-	$2 \cdot 10^{-1}$
3	$4 \cdot 10^{-1}$	$1.5 \cdot 10^{-1}$	-

## Flow diagram

16



## Sampling the time of transition

17

- The rate of transition of the system out of its current configuration (1, 1, 1) is:

$$\lambda^{(1,1,1)} = \lambda_{1 \rightarrow 2}^A + \lambda_{1 \rightarrow 3}^A + \lambda_{1 \rightarrow 2}^B + \lambda_{1 \rightarrow 3}^B + \lambda_{1 \rightarrow 2}^C + \lambda_{1 \rightarrow 3}^C$$

- We are now in the position of sampling the first system transition time  $t_1$ , by applying the **inverse transform method**:

$$t_1 = t_0 - \frac{1}{\lambda^{(1,1,1)}} \ln(1 - R_1)$$

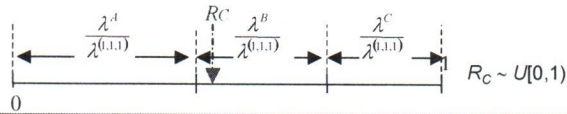
where  $R_1 \sim U[0,1)$

- Assuming that  $t_1 < T_M$  (otherwise we would proceed to the successive trial), we now need to determine which component has undergone the transition
- The probabilities of components A, B, C undergoing a transition out of their initial nominal states 1, given that a transition occurs at time  $t_1$ , are:

$$\frac{\lambda^A}{\lambda^{(1,1,1)}}, \quad \frac{\lambda^B}{\lambda^{(1,1,1)}}, \quad \frac{\lambda^C}{\lambda^{(1,1,1)}}$$

$$\lambda^A = \lambda_{1 \rightarrow 2}^A + \lambda_{1 \rightarrow 3}^A \quad \lambda^B = \lambda_{1 \rightarrow 2}^B + \lambda_{1 \rightarrow 3}^B \quad \lambda^C = \lambda_{1 \rightarrow 2}^C + \lambda_{1 \rightarrow 3}^C$$

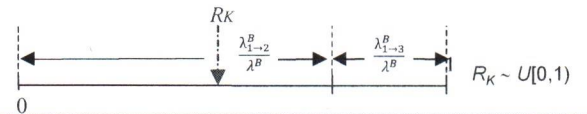
- Thus, we can apply the inverse transform method to the discrete distribution



- Since component B is the one undergoing the transition we need to sample the new state of component B.
- The probabilities of components B undergoing a transition out of their initial nominal states 1 given that a transition occurs at time  $t_1$ , are:

$$\frac{\lambda_{1 \rightarrow 2}^B}{\lambda^B}, \quad \frac{\lambda_{1 \rightarrow 3}^B}{\lambda^B}$$

- Thus, we can apply the inverse transform method to the discrete distribution



## Next step

- As a result of this first transition, at  $t_1$  the system is operating in configuration (1,2,1).
- The simulation now proceeds to sampling the next transition time  $t_2$  with the updated transition rate

$$\lambda^{(1,2,1)} = \lambda_{1 \rightarrow 2}^A + \lambda_{1 \rightarrow 3}^A + \lambda_{2 \rightarrow 3}^B + \lambda_{1 \rightarrow 2}^C + \lambda_{1 \rightarrow 3}^C$$